

Lecture11

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Lecture11

Linear Algebra

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(Department of CE)

Lecture #12

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Orthogonality

- Generalization of the linear structure (addition and scalar multiplication) of \mathbb{R}^2 and \mathbb{R}^2 leads the definition of a linear space.
- How to generalize the concepts of **length** and **angle**?
- These concepts are embedded in the concept we now investigate, **inner products**

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$\|x\| = \sqrt{x_1^2 + x_2^2}$

$\|x\|^2 + \|y\|^2 = \|x-y\|^2$

$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ $y = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$

$\sin \alpha = \frac{x_2}{\|x\|}$ $\cos \alpha = \frac{y_1}{\|y\|}$

$\sin \beta = \frac{x_1}{\|x\|}$ $\cos \beta = \frac{y_2}{\|y\|}$

$\theta = ?$ $\theta = \beta - \alpha$

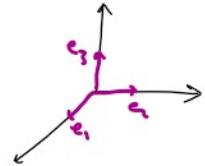
$\cos \theta = \cos(\beta - \alpha) = \cos \beta \cos \alpha + \sin \beta \sin \alpha$

$= \frac{x_1 y_1 + x_2 y_2}{\|x\| \|y\|}$

$\cos \theta = \frac{x_1 y_1 + x_2 y_2}{\|x\| \|y\|} = \frac{\langle x, y \rangle}{\|x\| \|y\|}$

Coordinate axes

- Every time we think of the $x - y$ plane or three-dimensional space or \mathbb{R}^n , the axes are there.
- They are usually perpendicular!
- The coordinate axes that the imagination constructs are practically always **orthogonal**.
- In choosing a basis, we tend to choose an orthogonal basis.



$$\langle x, y \rangle = 0$$

Inner products on real linear space

Definition:

An inner product on V is a function $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{R}$ such that

$$\|v\| = \sqrt{\langle v, v \rangle}$$

$$\textcircled{1} \langle v, v \rangle \geq 0 \text{ for all } v \in V.$$

مثبت بودن

$$\textcircled{2} \langle v, v \rangle = 0 \text{ if and only if } v = 0.$$

صفر بودن

$$\langle \cdot, \cdot \rangle : V \rightarrow \mathbb{R}$$

$$\textcircled{3} \langle u + v, w \rangle = \langle u, w \rangle + \langle v, w \rangle \text{ for all } u, v, w \in V.$$

خطی

$$\textcircled{4} \langle cu, w \rangle = c \langle u, w \rangle \text{ for all } u, w \in V \text{ and } c \in \mathbb{R}.$$

$$\textcircled{5} \langle v, w \rangle = \langle w, v \rangle.$$

تقارن

$$\langle u, v \rangle \in \mathbb{R}$$

$$\langle v, 0 \rangle = \langle 0, v \rangle = 0$$

$$\langle 0, v \rangle + \langle u, v \rangle = \langle u, v \rangle$$

$$\langle u, \cdot \rangle : V \rightarrow \mathbb{R}$$

$$\langle u, w + v \rangle = \langle u, w \rangle + \langle u, v \rangle$$

$$\langle u, cw \rangle = c \langle u, w \rangle$$

$$\langle w + v, u \rangle = \langle w, u \rangle + \langle v, u \rangle$$

Example

- The Euclidean inner product on \mathbb{R}^n :

$$\begin{cases} \langle \cdot, \cdot \rangle : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R} \\ \langle x, y \rangle = y^T x = y_1 x_1 + \cdots + y_n x_n. \end{cases}$$

$$\text{where } x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \text{ and } y = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix}.$$

$$\langle x, x \rangle = x_1^2 + \cdots + x_n^2 \geq 0$$

$$\langle x, x \rangle = 0 \iff x = 0$$

$$\begin{aligned} \langle x+z, y \rangle &= y^T (x+z) = y^T x + y^T z \\ &= \langle x, y \rangle + \langle z, y \rangle \end{aligned}$$

$$\langle cx, y \rangle = y^T (cx) = cy^T x$$

$$\begin{aligned} \langle x, y \rangle &= y^T x = \sum y_i x_i = x^T y \\ &= \langle y, x \rangle \end{aligned}$$

Example

- The linear space $P_n(x)$ of all polynomials with coefficients in \mathbb{R} and degree at most n :

$$\begin{cases} \langle \cdot, \cdot \rangle : P_n(x) \times P_n(x) \rightarrow \mathbb{R} \\ \langle f, g \rangle = \int_0^1 f(x)g(x)dx \end{cases}$$

$$\langle f+h, g \rangle$$

$$1 - \int_0^1 f(x)^2 dx \geq 0$$

$$2 - \int_0^1 f(x)^2 dx = 0 \iff f(x) = 0$$

$$3 - \int_0^1 (f(x)+h(x))g(x) dx = \int_0^1 fg dx + \int_0^1 hg dx$$

In an inner-product space:

- 1 In an inner-product space, we have additivity and the homogeneity in the second slot, as well as the first slot:

$$\langle u, v+w \rangle = \langle u, v \rangle + \langle u, w \rangle.$$

$$\langle u, cv \rangle = c \langle u, v \rangle.$$

- 2 For each fixed $u \in V$ the function

$$\langle u, - \rangle : V \rightarrow \mathbb{R}$$

is a linear map from $V \rightarrow \mathbb{R}$.

- 3 For each fixed $u \in V$,

$$\langle u, 0 \rangle = \langle 0, u \rangle = 0$$

Bilinear function

- Consider a linear function $T : \mathbb{R}^n \rightarrow \mathbb{R}$. Assuming the standard basis for \mathbb{R}^n , we have

$$T(x) = x_1 T(e_1) + \cdots + x_n T(e_n).$$

Let $y_i = T(e_i)$ for $1 \leq i \leq n$. Thus

$$T(x) = x_1 y_1 + \cdots + x_n y_n = y^T x.$$

- So each the linear function $T : \mathbb{R}^n \rightarrow \mathbb{R}$ is determined by a fixed vector $y \in \mathbb{R}^n$.
- Naturally, we can consider

$$\begin{cases} B : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R} \\ B(x, y) = y^T x \end{cases}$$

which is linear with respect to each of its variables.

Inner product as a bilinear functions

- The Euclidean inner product on \mathbb{R}^n , $\langle x, y \rangle$ is naturally defined as a bilinear function:

$$\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{R}$$

$$\langle u, - \rangle : V \rightarrow \mathbb{R}$$

$$\langle - , u \rangle : V \rightarrow \mathbb{R}$$

$$y \in \mathbb{R}^n$$

$$T(e_1) = y_1$$

$$\vdots$$

$$T(e_n) = y_n$$

$$\equiv B(-, y) : \mathbb{R}^n \rightarrow \mathbb{R}$$

$$T : \mathbb{R}^n \rightarrow \mathbb{R}$$

$$T(x) = y^T x = B(x, y)$$

$$B(x, y) : \mathbb{R}^n \rightarrow \mathbb{R}$$

$$\begin{aligned} * \langle v, v \rangle &\geq 0 \\ * \langle v, v \rangle &= 0 \iff v = 0 \\ - &\text{ خطرت } \\ - &\text{ متناهي } \end{aligned}$$

$$B : \mathbb{R}^n$$

$$\langle \cdot, \cdot \rangle : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$$

Inner product as a bilinear functions

- The Euclidean inner product on \mathbb{R}^n , $\langle x, y \rangle$ is naturally defined as a bilinear function:

$$\begin{cases} \langle \cdot, \cdot \rangle : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R} \\ \langle x, y \rangle = y^T x. \end{cases}$$

$$\begin{aligned} & \underbrace{\quad \quad \quad}_{\equiv} \\ & \mathbb{B} : \mathbb{R}^n \\ & \langle \cdot, \cdot \rangle : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R} \end{aligned}$$

Inner product spaces

Definition

An inner product space is a linear space with a binary operation called an **inner product**.

- The Euclidean inner product on \mathbb{R}^n with the inner product $\langle x, y \rangle = y^T x$ for each $x, y \in \mathbb{R}^n$.
- The linear space $P_n(x)$ of all polynomials with coefficients in \mathbb{R} and degree at most n with an inner product

$$\begin{cases} \langle \cdot, \cdot \rangle : P_n(x) \times P_n(x) \rightarrow \mathbb{R} \\ \langle f, g \rangle = \int_0^1 f(x)g(x)dx \end{cases}$$

is an inner product space.

$$\underline{\underline{\langle \cdot, \cdot \rangle : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}}}$$

$$\det : \underbrace{\mathbb{R} \times \dots \times \mathbb{R}}_{n \text{ times}} \rightarrow \mathbb{R}$$

$$T : V \rightarrow W$$

$$\underline{\underline{T : V \rightarrow \mathbb{R}^2}}$$

Norms

Definition

For $v \in V$, we define the norm of v , denoted $\|v\|$, by

$$\|v\| = \sqrt{\langle v, v \rangle}.$$

Examples

Example. With the Euclidean inner product:

$$\|x\| = \sqrt{x_1^2 + \dots + x_n^2}.$$

where $x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n$.

Example. With inner product on P_n :

$$\|f(x)\| = \sqrt{\int_0^1 f(x)^2 dx}$$

Orthonormal vectors in V

Definition

Two vectors $u, v \in V$ are said to be orthogonal if $\langle u, v \rangle = 0$.

Useful fact

If **nonzero vectors** v_1, \dots, v_n are **mutually orthogonal** (every vector is perpendicular to every other), then those vectors are linearly independent.

Why?

فرض : $\langle v_i, v_j \rangle = 0 \quad 1 \leq i, j \leq n$

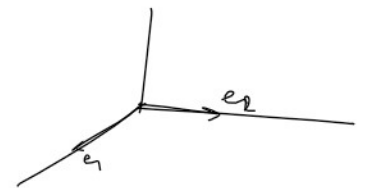
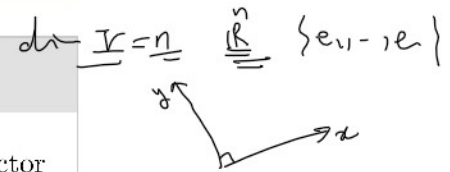
استلزام : استقلال متجهي

$$c_1 v_1 + \dots + c_n v_n = 0 \Rightarrow c_1 = \dots = c_n = 0$$

$$c_i = 0$$

$$0 = \langle c_1 v_1 + \dots + c_n v_n, v_i \rangle = \sum_{j=1}^n c_j \langle v_j, v_i \rangle$$

$$0 = c_i \langle v_i, v_i \rangle \quad , \quad \langle v_i, v_i \rangle \neq 0 \Rightarrow c_i = 0 \quad \forall i \quad 1 \leq i \leq n$$



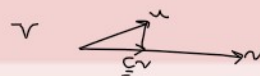
$$j \neq i \quad \langle v_j, v_i \rangle = 0$$

Cauchy-Schwarz Inequality

Proposition

Let V be a linear space and $u, v \in V$. Then

$$|\langle u, v \rangle| \leq \|u\| \|v\|$$



$$\langle u, v \rangle = c \langle v, v \rangle$$

$$\langle u, v \rangle = c \langle v, v \rangle$$

$$\langle u - cv, v \rangle = 0 \quad c = \frac{\langle u, v \rangle}{\langle v, v \rangle}$$

$$v \neq 0$$

$$p = ?$$

$$p = \frac{\langle u, v \rangle}{\langle v, v \rangle} v$$

$$u - p = u - \frac{\langle u, v \rangle}{\langle v, v \rangle} v$$

$$\|u\|^2 = \|p\|^2 + \|u - p\|^2 \geq \|p\|^2$$

$$p = \frac{\langle u, v \rangle}{\|v\|^2} v$$

$$\Rightarrow \|p\|^2 = \frac{\langle u, v \rangle^2}{\|v\|^4} \|v\|^2 = \frac{\langle u, v \rangle^2}{\|v\|^2}$$

$$\Rightarrow \|u\|^2 \geq \frac{\langle u, v \rangle^2}{\|v\|^2}$$

$v \neq 0$
 $\langle u - p, v \rangle = 0 \quad c = \frac{\langle u, v \rangle}{\langle v, v \rangle}$
 $p = ? \quad p = c v$
 $\langle u - p, v \rangle = 0$
 $0 = \langle u - p, v \rangle = v^T(u - p) = v^T u - v^T p = v^T u - v^T(c v) = v^T u - c \langle v, v \rangle$
 $\Rightarrow v^T u - c \langle v, v \rangle = 0 \Rightarrow c = \frac{v^T u}{\langle v, v \rangle} \Rightarrow p = \frac{v^T u}{\langle v, v \rangle} v$
 $u = p + (u - p)$
 $\|u\|^2 = \langle u, u \rangle = \langle p + (u - p), p + (u - p) \rangle = \langle p, p \rangle + \langle p, u - p \rangle + \langle u - p, p \rangle + \langle u - p, u - p \rangle$
 $\leq \|p\|^2 + \|u - p\|^2$

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$\Rightarrow \|u\|^2 \geq \frac{\langle u, v \rangle^2}{\langle v, v \rangle}$
 $\Rightarrow \|u\| \geq \frac{|\langle u, v \rangle|}{\|v\|}$
 $u = \frac{\langle u, v \rangle}{\langle v, v \rangle} v + (u - \frac{\langle u, v \rangle}{\langle v, v \rangle} v)$
 $v, u \in V \rightarrow$

$\langle u, v \rangle = \overline{\langle v, u \rangle}$
 $|\langle u, v \rangle| \leq \|u\| \|v\|$

Triangle Inequality

Proposition

Let V be a linear space. If $u, v \in V$, then

$$\|u + v\| \leq \|u\| + \|v\|$$

$\|u + v\|^2 = \langle u + v, u + v \rangle$
 $= \langle u, u \rangle + \langle u, v \rangle + \langle v, u \rangle + \langle v, v \rangle$
 $= \|u\|^2 + 2\langle u, v \rangle + \|v\|^2$
 $\leq \|u\|^2 + 2\|u\|\|v\| + \|v\|^2 = (\|u\| + \|v\|)^2$

Thank You!